L Number	Hits	Search Text	DB	Time stamp
4	19	(quer\$3 near4 (classif\$5 cluster\$3)) and (context\$1 near3 vector\$1)	USPAT; US-PGPUB; EPO; JPO; DERWENT;	2004/07/28 14:54
5	24	(quer\$3 near3 context\$1) same (context\$1 near3 vector\$1)	IBM_TDB USPAT; US-PGPUB; EPO; JPO;	2004/07/28 14:54
6	3	(quer\$3 near3 classif\$6) and (context\$1 near3 vector\$3)	DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO;	2004/07/28 14:55
7	1	US-6502091-B1.DID. and 6502091.PN. and (6502091.PN. and (6502091.pn. and (user\$4 same context\$4)))	DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT;	2004/07/28 14:55
8	24	(((analyz\$3 pars\$3) near5 quer\$3) and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3)	IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT;	2004/07/28 14:55
9	1	6502091.PN. and (6502091.PN. and (6502091.pn. and (user\$4 same context\$4))) and ((quer\$3 near3 context\$1) same (context\$1 near3 vector\$1))	IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT;	2004/07/28 14:55
10	0	6502091.PN. and (6502091.PN. and (6502091.pn. and (user\$4 same context\$4))) and ((((analyz\$3 pars\$3) near5 quer\$3) and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3))	IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT;	2004/07/28 14:56
11	13	((user\$3 near5 interact\$4 near5 (stat\$3 data information)) (user\$4 near5 (histor\$4 log\$4))) same (user\$4 near5 context\$4 near5 (vector\$4 classif\$4 cluster\$4))	IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT;	2004/07/28 14:56
12	0	(6502091.PN. and (6502091.pn. and (user\$4 same context\$4))) and ((((analyz\$3 pars\$3) near5 quer\$3)) and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3))	IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT;	2004/07/28 14:56
13	0	((((analyz\$3 pars\$3) near5 quer\$3) and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3)) and (6502091.PN. and (6502091.pn. and (user\$4 same context\$4))) and ((quer\$3 near3 context\$1)	IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT;	2004/07/28 14:56
14	0	same (context\$1 near3 vector\$1))) (6502091.PN. and (6502091.PN. and (6502091.pn. and (user\$4 same context\$4))) and ((quer\$3 near3 context\$1) same (context\$1 near3 vector\$1))) and ((((analyz\$3 pars\$3) near5 quer\$3) and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3)) and (((user\$3 near5 interact\$4 near5 (stat\$3 data information)) (user\$4 near5 (histor\$4 log\$4))) same (user\$4 near5 context\$4 near5	IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	2004/07/28 14:56
15	0	(vector\$4 classif\$4 cluster\$4))) ((((analyz\$3 pars\$3) near5 quer\$3) and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3)) and (((user\$3 near5 interact\$4 near5 (stat\$3 data information)) (user\$4 near5 (histor\$4 log\$4))) same (user\$4 near5 context\$4 near5 (vector\$4 classif\$4 cluster\$4))) and ((6502091.PN. and (6502091.pn. and (user\$4 same context\$4))) and ((((analyz\$3 pars\$3) near5 quer\$3)) and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3)))	USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	2004/07/28 14:56

16	0	(((user\$3 near5 interact\$4 near5 (stat\$3 data information))	USPAT;	2004/07/28 14:56
		(user\$4 near5 (histor\$4 log\$4))) same (user\$4 near5	US-PGPUB;	
		context\$4 near5 (vector\$4 classif\$4 cluster\$4))) and (EPO; JPO;	
		(6502091.PN. and (6502091.pn. and (user\$4 same	DERWENT;	
		context\$4))) and ((((analyz\$3 pars\$3) near5 quer\$3) and	IBM_TDB	
		((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3)		
		near5 quer\$3))) and (((((analyz\$3 pars\$3) near5 quer\$3))		
		and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4		
		cluste4\$3) near5 quer\$3)) and (6502091.PN. and		
		(6502091.PN. and (6502091.pn. and (user\$4 same		
	1	context\$4))) and ((quer\$3 near3 context\$1) same (context\$1		
		near3_vector\$1)))) and ((6502091.PN. and (6502091.PN. and		
		(6502091.pn. and (user\$4 same context\$4))) and ((quer\$3		
		near3 context\$1) same (context\$1 near3 vector\$1))) and		
	!	((((analyz\$3 pars\$3) near5 quer\$3) and ((histor\$3 log\$3)		
		near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3)) and		
		(((user\$3 near5 interact\$4 near5 (stat\$3 data information))		
		(user\$4 near5 (histor\$4 log\$4))) same (user\$4 near5		
		context\$4 near5 (vector\$4 classif\$4 cluster\$4))))		
17	0	((6502091.PN, and (6502091.pn, and (user\$4 same	USPAT;	2004/07/28 14:56
		context\$4))) and ((((analyz\$3 pars\$3) near5 quer\$3) and	US-PGPUB:	
		((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4 cluste4\$3)	EPO; JPO;	
		near5 quer\$3))) and (((((analyz\$3 pars\$3) near5 quer\$3))	DERWENT;	
		and ((histor\$3 log\$3) near5 quer\$3)) and ((classif\$4	IBM TDB	
		cluste4\$3) near5_quer\$3)) and (6502091.PN. and	_	
		(6502091.PN. and (6502091.pn. and (user\$4 same		
		context\$4))) and ((quer\$3 near3 context\$1) same (context\$1		
		near3 vector\$1)))) and ((6502091.PN. and (6502091.PN. and		
		(6502091.pn. and (user\$4 same context\$4))) and ((quer\$3		
		near3 context\$1) same (context\$1 near3 vector\$1))) and		
		((((analyz\$3 pars\$3) near5 quer\$3) and ((histor\$3 log\$3)		
		near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3)) and		
		(((user\$3 near5 interact\$4 near5 (stat\$3 data information))		
		(user\$4 near5 (histor\$4 log\$4))) same (user\$4 near5		
,		context\$4 near5 (vector\$4 classif\$4 cluster\$4))))		
-	19	context\$1 near4 attribut\$3 near4 database\$1	USPAT;	2004/02/25 10:17
			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	641	context\$3 near2 vector\$3	USPAT;	2003/02/11 08:09
			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
	_	500004	IBM_TDB	
-	2	5303361.pn.	USPAT;	2004/02/25 13:47
			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
_	2	5321833.pn.	IBM_TDB USPAT;	2003/02/11 10:52
	2	, 552 1555.pm.	US-PGPUB;	2003/02/11 10.32
			EPO; JPO;	
			DERWENT:	
			IBM_TDB	
<u>-</u>	2	5524187.pn.	USPAT;	2003/02/11 10:53
	*	σοετιστ.μπ.	US-PGPUB;	2003/02/11 10.53
			EPO; JPO;	
		*	DERWENT;	
			IBM_TDB	
_	2	5546516.pn.	USPAT;	2003/02/11 10:53
		об 100 го.ргі.	US-PGPUB;	2003/02/11 10.53
	1		EPO; JPO;	
			DERWENT:	
			IBM_TDB	
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Perc. yPoth Derwent	10:54
DERWENT; IBM, TDB USPAT; US-PGPUB; EPG; JPG; DERWENT; IBM, TDB USPAT; US-PGPUB; EPG; DERWENT; IBM, TDB USPAT; US-PGPUB; EPG; DERWENT; IBM, TDB USPAT; US-PGPUB; EPG; DERWENT; IBM, TDB USPAT; US-PGPUB;	
BM TDB	
- 2 5608899.pn. USPAT: USP-GPUB: EPO; JPO; DERWENT: IBM TDB USPAT:	
US-PGPUB; EPO; JPO; DERWENT; IBM_TDB	
EPO; JPO; DERWENT; IBM, TDB USPAT; US.PGPUB; EPO; JPO; DERWENT; IB	10:54
DERWENT: IBM_TDB USPAT: US_PGPUB; EPO; JPO; DERWENT: IBM_TDB USP	
BM TDB	
- 2 5619709.pn.	
US-PGPUB EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB EPO; JPO;	
EPC, JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; I	10:54
- 2 5710899.pn. DERWENT; IBM. TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM. TDB USPAT; US-PGPUB; US-PGPU	
- 2 5710899.pn.	
- 2 5710899.pn. USPAT; US-PGPUB; PO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; US-PGPUB; EPO; JPO; DERWENT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB US-PGPUB; EPO; JPO; DERWE	
- 2 5754939.pn. - 2 5754939.pn. - 2 5768578.pn. - 2 5768578.pn. - 2 5930501.pn. - 2 5826260.pn. - 2 5846891.pn. - 3 quer\$3 near4 context\$1 near4 classif\$6 - 715 quer\$3 near4 context\$1 - 2 12 5754939.pn. - 2 2 5754939.pn. - 3 Quo3/02/11 - 2 2 5768578.pn. - 3 Quo3/02/11 - 2 2 5768578.pn. - 3 Quo3/02/11 - 2 2 5768578.pn. - 3 Quo3/02/11 - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 Quo3/02/11 - 5754939.pn. - 2 2 5768578.pn. - 2 2 5768578.pn. - 3 Quo3/02/11 - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 2 5826260.pn. - 4 2 5826260.pn. - 5768578.pn. - 2 2 5826260.pn. - 2 2 5826260.pn. - 3 Quo3/02/11 - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 2 5826260.pn. - 4 2 5826260.pn. - 5768578.pn. - 2 2 5826260.pn. - 3 Quo3/02/11 - 2 2 5826260.pn. - 4 2 5826260.pn. - 5 2 5826260.pn. - 5 2 5826260.pn. - 5 2 5826260.pn. - 5 2 5826260.pn. - 2 2 5826260.pn. - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 2 5826260.pn. - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 2 5826260.pn. - 5 2 5826260.pn. - 5 2 5826260.pn. - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 2 5826260.pn. - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 2 5826260.pn. - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 2 5826260.pn. - 2 2 5826260.pn. - 3 Quo3/02/11 - 4 2 5826260.pn. - 4 2 5826260.pn. - 5 2 5826260.pn. - 5 2 5826260.pn. - 5 2 5826260.pn. - 2 2 5826260.pn. - 2 2 5826260.pn. - 3 2 2 5826260.pn. - 4 2 5826260.pn. - 5 2 5826260.pn. - 5 2 5826260.pn. - 5 2 5826260.pn. - 2 2 5826260.pn. - 5 2 5826260.pn. - 6 2 5 5 5 5 5 5 5 5 5	
- 2 5754939.pn. - 2 5754939.pn. - 2 5768578.pn. - 2 5768578.pn. - 2 5930501.pn. - 2 5930501.pn. - 2 5826260.pn. - 2 5826260.pn. - 2 5446891.pn. - 3 quer\$3 near4 context\$1 near4 classif\$6 - 715 quer\$3 near4 context\$1 - 2 5754939.pn. - 2 5768578.pn. - 2 5768578.pn. - 2 5768578.pn. - 3 2 5768578.pn. - 2 5930501.pn. - 3 2 5930501.pn. - 4 2 5930501.pn. - 5	10:55
- 2 5754939.pn. - 2 5754939.pn. - 2 5768578.pn. - 2 5768578.pn. - 2 5930501.pn. - 2 5826260.pn. - 2 5826260.pn. - 3 quer\$3 near4 context\$1 near4 classif\$6 - 3 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1 - 2 5754939.pn. - 3 Q003/02/11 - 1 2 0003/02/11 - 2 0003/02/11 - 2 0003/02/11 - 3 Q003/02/11	
BIM_TDB USPAT; US-PGPUB; EPC; JPC; DERWENT; IBM_TDB USPAT; US-	
- 2 5764939.pn. USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT;	
US-PGPUB; EPO; JPO; DERWENT; IBM TDB USPAT; US-PGPUB; EPO; JPO; DERWENT, IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT, IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DE	
EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM	10:55
DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM	
BM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; US-PGPUB; US-PGP	
- 2 5768578.pn. USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT;	
US-PGPUB; EPO; IPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; IPO; DERWENT; IPO; DE	
EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; I	10:56
- 2 5930501.pn.	
- 2 5930501.pn. IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_	
- 2 5930501.pn. USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; USP	
US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DE	
EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM	111:02
DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT	
- 2 5826260.pn. IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; ISPAT; US	
- 2 5826260.pn. USPĀT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; US-PGPUB; EPO; JPO; DERWENT; US-PGPUB; EPO; JPO; DERWENT;	
- 2 5446891.pn. US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; US-PGPUB	
EPO; JPO; DERWENT; IBM_TDB US-PAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; USPAT; US-PGPUB; USPAT; US-PGPUB; USPAT; US-PGPUB; US	11:03
- 2 5446891.pn. DERWENT; IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; US-PGPUB; EPO; DERWENT; US-PGPUB;	
- 2 5446891.pn. IBM_TDB USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB;	
- 2 5446891.pn. USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; USPAT; US-PGPUB; USPAT; US-PGPUB;	
- 26 quer\$3 near7 (context\$3 near2 vector\$3) - 26 quer\$3 near7 (context\$3 near2 vector\$3) - 3 quer\$3 near4 context\$1 near4 classif\$6 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1	
- 26 quer\$3 near7 (context\$3 near2 vector\$3) - 26 quer\$3 near7 (context\$3 near2 vector\$3) - 3 quer\$3 near4 context\$1 near4 classif\$6 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1	I 11:03
- 26 quer\$3 near7 (context\$3 near2 vector\$3) - 26 quer\$3 near7 (context\$3 near2 vector\$3) - 3 quer\$3 near4 context\$1 near4 classif\$6 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1	
- 26 quer\$3 near7 (context\$3 near2 vector\$3) - 3 quer\$3 near4 context\$1 near4 classif\$6 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1	
- 26 quer\$3 near7 (context\$3 near2 vector\$3) - 3 quer\$3 near4 context\$1 near4 classif\$6 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1	
- 3 quer\$3 near4 context\$1 near4 classif\$6 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1 - 715 quer\$3 near4 context\$1	
EPO; JPO; DERWENT; IBM_TDB uspat; Us-PGPUB; EPO; JPO; DERWENT; IBM_TDB uspat; Us-PGPUB; EPO; JPO; DERWENT; IBM_TDB uspat; Uspat; IBM_TDB uspat; Uspat	3 10:32
- 3 quer\$3 near4 context\$1 near4 classif\$6 USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB - 715 quer\$3 near4 context\$1 USPAT; USPAT	
- 3 quer\$3 near4 context\$1 near4 classif\$6 USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; Quer\$3 near4 context\$1 USPAT; USPA	
- 3 quer\$3 near4 context\$1 near4 classif\$6 USPAT; US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; US-PGPUB; USPAT; US-PGPUB; USPAT; US-PGPUB;	
US-PGPUB; EPO; JPO; DERWENT; IBM_TDB USPAT; 2003/07/16 US-PGPUB; US-PGPUB; US-PGPUB;	
EPO; JPO; DERWENT; IBM_TDB - 715 quer\$3 near4 context\$1 USPAT; US-PGPUB;	3 10:34
DERWENT; IBM_TDB USPAT; 2003/07/16 US-PGPUB;	
- 715 quer\$3 near4 context\$1 IBM_TDB USPAT; 2003/07/16 US-PGPUB;	
- 715 quer\$3 near4 context\$1 USPAT; 2003/07/16	
US-PGPUB;	
	3 10:39
EPO; JPO;	
DERWENT;	
IBM_TDB	
- 599 quer\$3 near3 context\$1 USPAT; 2003/07/16	3 10:35
US-PGPUB;	
EPO; JPO;	
DERWENT;	
IBM_TDB	

-	20	(quer\$3 near3 context\$1) same (context\$1 near3 vector\$1)	USPAT; US-PGPUB;	2004/07/28 14:54
			EPO; JPO; DERWENT;	
			IBM_TDB	
-	27	(quer\$3 near3 context\$1) and (context\$1 near3 vector\$1)	USPAT;	2003/07/16 10:55
			US-PGPUB; EPO; JPO;	0
			DERWENT;	
			IBM_TDB	
-	16	quer\$3 near3 context\$1 near3 vector\$1	USPAT;	2003/07/16 10:42
			US-PGPUB; EPO; JPO;	
			DERWENT;	
		0 1 :00	IBM_TDB	0000/07//0 :- :-
-	165	quer\$3 near3 classif\$6	USPAT;	2003/07/16 10:42
			US-PGPUB; EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	3	(quer\$3 near3 classif\$6) and (context\$1 near3 vector\$3)	USPAT;	2004/07/28 14:54
			US-PGPUB; EPO; JPO;	
			DERWENT;	
	0.40	intelline of the control of the cont	IBM_TDB	0000/07/10 10 11
-	342	intelligen\$4 near3 quer\$3	USPAT; US-PGPUB;	2003/07/16 10:44
			EPO; JPO;	
			DERWENT;	
		(intelligent Among and the little (04 and 15)	IBM_TDB	0000/07/40 40 45
_	9	(intelligen\$4 near3 quer\$3) and (context\$1 near3 vector\$3)	USPAT; US-PGPUB;	2003/07/16 10:45
			EPO; JPO;	
			DERWENT;	
	200	quart2 nagr4 (contactt2 many2 ventort2)	IBM_TDB	2002/07/46 40:40
-	26	quer\$3 near4 (context\$3 near2 vector\$3)	USPAT; US-PGPUB;	2003/07/16 10:46
			EPO; JPO;	
			DERWENT;	
_	502	quer\$4 near4 (classif\$5 cluster\$3)	IBM_TDB USPAT;	2003/07/16 10:55
_	302	queign neain (classifips clustelips)	US-PGPUB;	2003/07/10 10:00
			EPO; JPO;	
			DERWENT;	
_	14	(quer\$4 near4 (classif\$5 cluster\$3)) and (context\$1 near3	IBM_TDB USPAT;	2004/07/28 14:54
	1-4	vector\$1)	US-PGPUB;	2007/0//20 14:04
		· '	EPO; JPO;	
		,	DERWENT;	
_	53	(quer\$4 near4 (classif\$5 cluster\$3)) same vector\$1	IBM_TDB USPAT;	2003/07/16 11:04
		(425.4 - Float - (510001140 Glastor40)) dallie vootor41	US-PGPUB;	2000/07/10 11:04
			EPO; JPO;	
			DERWENT; IBM_TDB	
_	19	context\$1 near3 attribut\$3 near3 database\$1	USPAT;	2003/07/16 15:45
			US-PGPUB;	
		·	EPO; JPO;	
			DERWENT; IBM_TDB	
-	0	6327590.pn and (context\$1 near5 vector\$1)	USPAT;	2003/07/25 09:22
		·	US-PGPUB;	
			EPO; JPO; DERWENT;	
			IBM_TDB	
		<u> </u>		لــــــــــــــــــــــــــــــــــــ

-	0	6327590.pn and (context\$1 near10 vector\$1)	USPAT;	2003/07/25 09:21
			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	0	6327590.pn and (context\$1 same vector\$1)	USPAT;	2003/07/25 09:21
			US-PGPUB;	
			EPO; JPO;	
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	_		IBM_TDB	
-	0	6327590.pn and (context\$1 and vector\$1)	USPAT;	2003/07/25 09:22
			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	1	6327590.pn. and (context\$1 near5 vector\$1)	USPAT;	2003/07/25 09:22
	}		US-PGPUB;	
			EPO; JPO;	
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-	1	6327590.pn. and (context\$1 near10 vector\$1)	USPAT;	2003/07/25 09:22
1			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
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-	1	6327590.pn. and (context\$1 same vector\$1)	USPAT;	2003/07/25 10:34
			US-PGPUB;	
			EPO; JPO;	
l			DERWENT;	
	}		IBM_TDB	
-	293	user near3 quer\$3 near4 record\$1	USPAT;	2003/07/25 10:35
		*	US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	36	(user near3 quer\$3 near4 record\$1) and ((analyz\$4 pars\$3)	USPAT;	2003/07/25 10:36
		near4 record\$1)	US-PGPUB;	
		•	EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	37	(user near3 quer\$3 near4 record\$1) and ((analyz\$4 pars\$3)	USPAT;	2003/07/25 10:41
		near4 record\$3)	US-PGPUB;	
			EPO; JPO;	
		·	DERWENT;	
			IBM_TDB	
-	2071	(analyz\$4 pars\$3) near5 quer\$3	USPAT;	2003/07/25 10:59
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			EPO; JPO;	
			DERWENT;	
	1		IBM_TDB	
-	385	((analyz\$4 pars\$3) near5 quer\$3) and ((histor\$3 log\$3) near5	USPAT;	2003/07/25 10:48
		quer\$3)	US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	18	(((analyz\$4 pars\$3) near5 quer\$3) and ((histor\$3 log\$3)	USPAT;	2004/07/28 14:55
		near5 quer\$3)) and ((classif\$4 cluste4\$3) near5 quer\$3)	US-PGPUB;	
		·	EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	2096	record\$3 near3 quer\$3	USPAT;	2003/07/25 10:48
			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
L			IBM_TDB	

-	1479	record\$3 near2 quer\$3	USPAT;	2003/07/25 10:48
			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
	000	(IBM_TDB	2002/07/25 10:40
-	266	(record\$3 near2 quer\$3) and ((analyz\$4 pars\$3) near5	USPAT; US-PGPUB;	2003/07/25 10:49
		quer\$3)	EPO; JPO;	
			DERWENT;	
			IBM_TDB	
_	109	((record\$3 near2 quer\$3) and ((analyz\$4 pars\$3) near5	USPAT;	2003/07/25 10:54
		quer\$3)) and (classif\$4 cluster\$3)	US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	42	quer\$3 near3 log\$1 near3 file\$1	USPAT;	2003/07/25 10:59
			US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
	4047		IBM_TDB	2002/07/25 10:50
-	1247	usag\$3 near3 log\$3	USPAT; US-PGPUB;	2003/07/25 10:59
			EPO; JPO;	
			DERWENT;	
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_	11	(usag\$3 near3 log\$3) and ((analyz\$4 pars\$3) near5 quer\$3)	USPAT;	2003/07/25 11:00
	1	(dodgto nodro logto) and ((disci)24 : paroto) nodro querto)	US-PGPUB;	
			EPO; JPO;	
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_	47991	(user\$4 near5 interact\$4 near5 (stat\$3 data information))	USPAT;	2004/02/24 15:56
		(user\$4 near5 (histor\$4 log\$4))	US-PGPUB;	
			EPO; JPO;	
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	12	((uportA moorE interacttA nearE (statt2 data information))	IBM_TDB	2004/07/28 14:55
-	12	((user\$4 near5 interact\$4 near5 (stat\$3 data information)) (user\$4 near5 (histor\$4 log\$4))) same (user\$4 near5	USPAT; US-PGPUB;	2004/07/20 14.55
		context\$4 near5 (vector\$4 classif\$4 cluster\$4))	EPO; JPO;	*
		Contextor Hears (vectoror classifor clasteror))	DERWENT;	
			IBM_TDB	
_	789	(user\$4 near5 quer\$4 near5 (histor\$4 log\$4))	USPAT;	2004/02/24 16:01
		(,	US-PGPUB;	
			EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	3		USPAT;	2004/02/24 15:58
		(user\$4 near5 context\$4 near5 (vector\$4 classif\$4 cluster\$4))	US-PGPUB;	
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			DERWENT; IBM_TDB	
	3731	(guer\$4 near5 (histor\$4 log\$4))	USPAT;	2004/02/24 15:58
-	3/31	(quer\$4 fleat5 (flistor\$4 log\$4))	US-PGPUB:	2004/02/24 15.50
			EPO; JPO;	
			DERWENT;	
			IBM TDB	
_	5	((quer\$4 near5 (histor\$4 log\$4))) same (user\$4 near5	USPAT;	2004/02/24 16:00
		context\$4 near5 (vector\$4 classif\$4 cluster\$4))	US-PGPUB;	
		, , , , , , , , , , , , , , , , , , , ,	EPO; JPO;	
			DERWENT;	
			IBM_TDB	
-	15	((quer\$4 near5 (histor\$4 log\$4))) and (user\$4 near5	USPAT;	2004/02/24 16:09
		context\$4 near5 (vector\$4 classif\$4 cluster\$4))	US-PGPUB;	
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-	. 8	((quer\$4 near5 (histor\$4 log\$4))) same (context\$4 near5 (vector\$4 classif\$4 cluster\$4))	USPAT; US-PGPUB;	2004/02/24 16:02
		(VECIOI\$4 Classif4 Cluster\$4))	EPO; JPO;	
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- '	40987	(user\$4 near5 (histor\$4 log\$4))	USPAT;	2004/02/24 16:01
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-	12	((user\$4 near5 (histor\$4 log\$4))) same (context\$4 near5	USPAT;	2004/02/24 16:02
		(vector\$4 classif\$4 cluster\$4))	US-PGPUB;	
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	3	"09/778146"	IBM_TDB USPAT;	2004/02/24 16:09
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_	1	6456978.pn. AND (receiv\$4 same quer\$4)	USPAT;	2004/02/24 16:15
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			IBM_TDB	
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			US-PGPUB;	
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-	'	(6456978.pn. AND (receiv\$4 same quer\$4)) and (6456978.pn. AND (context\$4 samer vector\$4))	US-PGPUB;	2004/02/24 16:16
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<u>-</u>	1	6456978.pn. AND (user\$4 near10 context\$4)	USPAT;	2004/02/24 16:17
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-	1	((6456978.pn. AND (context\$4 same vector\$4)) and	USPAT;	2004/02/24 16:18
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-	0	6456978.pn. AND (context\$4 near5 attribut\$4)	USPAT; US-PGPUB:	2004/02/24 16:18
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-	1	6456978.pn. AND (context\$4 same attribut\$4)	USPAT;	2004/02/25 09:59
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-	24	(user\$3 near5 quer\$4 near5 histor\$4) and (user\$3 near5	USPAT;	2004/02/25 10:18
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_	1772	1or (user\$3 near5 quer\$4 near5 analy\$4)	USPAT:	2004/02/25 10:18
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	1	(classif\$4 cluster\$4))	US-PGPUB;	
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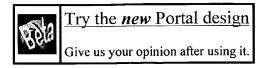
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_	3003030	bays et al	US-PGPUB;	2004/02/23 13.32
			EPO; JPO;	
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-	8646	search\$4 near5 (histor\$4 log\$4)	USPAT;	2004/02/25 13:53
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-	2	(((search\$4 near5 (histor\$4 log\$4)) same (classif\$4 cluster\$3))	USPAT;	2004/02/25 13:54
		and (user\$3 near5 context\$4)) and vector\$3	US-PGPUB;	
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_	15	((search\$4 near5 /histor\$4 log\$4)) same (alossif\$4 alustar\$2))	IBM_TDB USPAT;	2004/02/25 42:54
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-	2	6502091.pn. and (user\$4 same context\$4)	USPAT;	2004/02/25 14:21
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-	1	supervis\$4 near5 lean\$4 near5 algorithm\$3	USPAT;	2004/02/25 14:22
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			IBM_TDB	



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Search Results

Search Results for: [user <and> context <and> database <and> classifier <and> context <and> vector <and> interaction <and> history]
Found 45 of 139,988 searched.

Technique for automatically correcting words in text

87

Karen Kukich

ACM Computing Surveys (CSUR) December 1992

Volume 24 Issue 4

Research aimed at correcting words in text has focused on three progressively more difficult problems:(1) nonword error detection; (2) isolated-word error correction; and (3) context-dependent work correction. In response to the first problem, efficient pattern-matching and n-gram analysis techniques have been developed for detecting strings that do not appear in a given word list. In response to the second problem, a variety of general and application-specific spelling cor ...

2 Learning classifiers: Liveclassifier: creating hierarchical text classifiers through web 80 corpora

Chien-Chung Huang, Shui-Lung Chuang, Lee-Feng Chien

Proceedings of the 13th conference on World Wide Web May 2004

Many Web information services utilize techniques of information extraction(IE) to collect important facts from the Web. To create more advanced services, one possible method is to discover thematic information from the collected facts through text classification. However, most conventional text classification techniques rely on manual-labelled corpora and are thus ill-suited to cooperate with Web information services with open domains. In this work, we present a system named LiveClassifier that ...

Ontological user profiling in recommender systems

Stuart E. Middleton , Nigel R. Shadbolt , David C. De Roure

ACM Transactions on Information Systems (TOIS) January 2004

Volume 22 Issue 1

We explore a novel ontological approach to user profiling within recommender systems, working on the problem of recommending on-line academic research papers. Our two experimental systems, Quickstep and Foxtrot, create user profiles from unobtrusively monitored behaviour and relevance feedback, representing the profiles in terms of a research paper topic ontology. A novel profile visualization approach is taken to acquire profile feedback. Research papers are classified using ontological classes ...

0,

4 A model of multimedia information retrieval

Carlo Meghini , Fabrizio Sebastiani , Umberto Straccia

Journal of the ACM (JACM) September 2001

Volume 48 Issue 5

Research on multimedia information retrieval (MIR) has recently witnessed a booming interest. A prominent feature of this research trend is its simultaneous but independent materialization within several fields of computer science. The resulting richness of paradigms, methods and systems may, on the long run, result in a fragmentation of efforts and slow down progress. The primary goal of this study is to promote an integration of methods and techniques for MIR by contributing a conceptual model ...

5 Data clustering: a review

80

A. K. Jain , M. N. Murty , P. J. Flynn

ACM Computing Surveys (CSUR) September 1999

Volume 31 Issue 3

Clustering is the unsupervised classification of patterns (observations, data items, or feature vectors) into groups (clusters). The clustering problem has been addressed in many contexts and by researchers in many disciplines; this reflects its broad appeal and usefulness as one of the steps in exploratory data analysis. However, clustering is a difficult problem combinatorially, and differences in assumptions and contexts in different communities has made the transfer of useful generic co ...

6 Knowledge and representation: Leveraging a common representation for personalized search and summarization in a medical digital library

80

Kathleen R. McKeown , Noemie Elhadad , Vasileios Hatzivassiloglou

Proceedings of the third ACM/IEEE-CS joint conference on Digital libraries May 2003

Despite the large amount of online medical literature, it can be difficult for clinicians to find relevant information at the point of patient care. In this paper, we present techniques to personalize the results of search, making use of the online patient record as a sophisticated, pre-existing user model. Our work in *PERSIVAL*, a medical digital library, includes methods for re-ranking the results of search to prioritize those that better match the patient record. It also generates summa ...

7 Supporting cooperative and personal surfing with a desktop assistant

80

Hannes Marais , Krishna Bharat

Proceedings of the 10th annual ACM symposium on User interface software and technology October 1997

8 A multilevel approach to intelligent information filtering: model, system, and

80

ৰী evaluation

J. Mostafa , S. Mukhopadhyay , M. Palakal , W. Lam

ACM Transactions on Information Systems (TOIS) October 1997

Volume 15 Issue 4

In information-filtering environments, uncertainties associated with changing interests of the user and the dynamic document stream must be handled efficiently. In this article, a filtering model is proposed that decomposes the overall task into subsystem functionalities and highlights the need for multiple adaptation techniques to cope with uncertainties. A filtering system, SIFTER, has been implemented based on the model, using established techniques in information retrieval and artificia ...

9 System section: Computer vision techniques for PDA accessibility of in-house video 77

a surveillance

Rita Cucchiara, Costantino Grana, Andrea Prati, Roberto Vezzani

First ACM SIGMM international workshop on Video surveillance November 2003

In this paper we propose an approach to indoor environment surveillance and, in particular, to people behaviour control in home automation context. The reference application is a silent and automatic

control of the behaviour of people living alone in the house and specially conceived for people with limited autonomy (e.g., elders or disabled people). The aim is to detect dangerous events (such as a person falling down) and to react to these events by establishing a remote connection with low-per ...

10 Semantic annotation and integration: Towards the self-annotating web

77

Philipp Cimiano , Siegfried Handschuh , Steffen Staab

Proceedings of the 13th conference on World Wide Web May 2004

The success of the Semantic Web depends on the availability of ontologies as well as on the proliferation of web pages annotated with metadata conforming to these ontologies. Thus, a crucial question is where to acquire these metadata from. In this paper wepropose PANKOW (Pattern-based Annotation through Knowledge on theWeb), a method which employs an unsupervised, pattern-based approach to categorize instances with regard to an ontology. The approach is evaluated against the manual annotations ...

11 Message classification in the call center

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Stephan Busemann, Sven Schmeier, Roman G. Arens

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12 Special issue on word sense disambiguation: Introduction to the special issue on

77

word sense disambiguation: the state of the art

Nancy Ide , Jean Véronis

Computational Linguistics March 1998

Volume 24 Issue 1

13 Evaluating message understanding systems: an analysis of the third message

77

understanding conference (MUC-3)

Nancy Chinchor , David D. Lewis , Lynette Hirschman

Computational Linguistics September 1993

Volume 19 Issue 3

This paper describes and analyzes the results of the Third Message Understanding Conference (MUC-3). It reviews the purpose, history, and methodology of the conference, summarizes the participating systems, discusses issues of measuring system effectiveness, describes the linguistic phenomena tests, and provides a critical look at the evaluation in terms of the lessons learned. One of the common problems with evaluations is that the statistical significance of the results is unknown. In the disc ...

14 Dialogue act modeling for automatic tagging and recognition of conversational

77

त्री speech

Andreas Stolcke, Noah Coccaro, Rebecca Bates, Paul Taylor, Carol Van Ess-Dykema, Klaus Ries, Elizabeth Shriberg, Daniel Jurafsky, Rachel Martin, Marie Meteer

Computational Linguistics September 2000

Volume 26 Issue 3

We describe a statistical approach for modeling dialogue acts in conversational speech, i.e., speech-act-like units such as STATEMENT, QUESTION, BACKCHANNEL, AGREEMENT, DISAGREEMENT, and APOLOGY. Our model detects and predicts dialogue acts based on lexical, collocational, and prosodic cues, as well as on the discourse coherence of the dialogue act sequence. The dialogue model is based on treating the discourse structure of a conversation as a hidden ...

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Proceedings of the 2004 ACM symposium on Applied computing March 2004

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16 Special issue on learning from imbalanced datasets: Mining with rarity: a unifying

77

framework
Gary M. Weiss

ACM SIGKDD Explorations Newsletter June 2004

Volume 6 Issue 1

Rare objects are often of great interest and great value. Until recently, however, rarity has not received much attention in the context of data mining. Now, as increasingly complex real-world problems are addressed, rarity, and the related problem of imbalanced data, are taking center stage. This article discusses the role that rare classes and rare cases play in data mining. The problems that can result from these two forms of rarity are described in detail, as are methods for addressing these ...

17 Maximum likelihood estimation for filtering thresholds

77

Yi Zhang , Jamie Callan

Proceedings of the 24th annual international ACM SIGIR conference on Research and development in information retrieval September 2001

Information filtering systems based on statistical retrieval models usually compute a numeric score indicating how well each document matches each profile. Documents with scores above profile-specific dissemination thresholds are delivered. An optimal dissemination threshold is one that maximizes a given utility function based on the distributions of the scores of relevant and non-relevant documents. The parameters of the distribution can be estimated using releva ...

18 Reports from related meetings: Interface '99: a data mining overview

77

Arnold Goodman

ACM SIGKDD Explorations Newsletter January 2000

Volume 1 Issue 2

This personal overview of Interface '99 is intended to communicate its meaning and relevance to SIGKDD, as well as provide valuable information on trends within the Interface for data miners seeking to learn more about statistics. In addition, it is the newest link in a bridge between the Interface and KDD begun by References 2-4 and the sessions on KDD at Interface '98 and Interface '99.

19 Survey articles: Web usage mining: discovery and applications of usage patterns

77

from Web data

Jaideep Srivastava , Robert Cooley , Mukund Deshpande , Pang-Ning Tan

ACM SIGKDD Explorations Newsletter January 2000

Volume 1 Issue 2

Web usage mining is the application of data mining techniques to discover usage patterns from Web data, in order to understand and better serve the needs of Web-based applications. Web usage mining consists of three phases, namely *preprocessing*, *pattern discovery*, and *pattern analysis*. This paper describes each of these phases in detail. Given its application potential, Web usage mining has seen a rapid increase in interest, from both the research and practice communities. This pap ...

20 Visualization: Analysis of visualisation requirements for fuzzy systems

77

Binh Pham , Ross Brown

Proceedings of the 1st international conference on Computer graphics and interactive techniques in Austalasia and South East Asia February 2003

This paper provides a comprehensive analysis of the working and requirements of fuzzy systems with the view to devise appropriate visualisation framework and techniques for these systems using a user- and task-oriented approach. We firstly discuss the nature of fuzzy data and the essential components of typical fuzzy systems, then categorise different visualisation requirements from three perspectives: user of fuzzy systems, designer of fuzzy systems and designer of visualisation systems. The vi ...

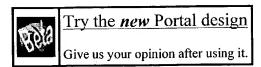
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21 Accepted Posters: Beyond broadcast

Kevin Livingston , Mark Dredze , Kristian Hammond , Larry Birnbaum

Proceedings of the 8th international conference on Intelligent user interfaces January 2003

The work presented in this paper takes a novel approach to the task of providing information to viewers of broadcast news. Instead of considering the broadcast news as the end product, this work uses it as a starting point to dynamically build an information space for the user to explore. This information space is designed to satisfy the users information needs, by containing more breadth, depth, and points of view than the original broadcast story. The architecture and current implementation ar ...

22 Papers: collaborating through documents: Augmenting shared personal calendars

Joe Tullio , Jeremy Goecks , Elizabeth D. Mynatt , David H. Nguyen

Proceedings of the 15th annual ACM symposium on User interface software and technology October 2002

In this paper, we describe Augur, a groupware calendar system to support personal calendaring practices, informal workplace communication, and the socio-technical evolution of the calendar system within a workgroup. Successful design and deployment of groupware calendar systems have been shown to depend on several converging, interacting perspectives. We describe calendar-based work practices as viewed from these perspectives, and present the Augur system in support of them. Augur allows users t ...

23 Poster session: Automated learning of model classifications
Cheuk Yiu Ip, William C. Regli, Leonard Sieger, Ali Shokoufandeh

Proceedings of the eighth ACM symposium on Solid modeling and applications June 2003
This paper describes a new approach to automate the classification of solid models using machine learning techniques. Existing approaches, based on group technology, fixed matching algorithms or pre-defined feature sets, impose a priori categorization schemes on engineering data or require significant human labeling of design data. This paper describes a shape learning algorithm and a general technique for "teaching" the algorithm to identify new or hidden classifications that are relevant in ma ...

77

77

Proceedings of the tenth international conference on World Wide Web April 2001

29 Temporal sequence learning and data reduction for anomaly detection

Terran Lane, Carla E. Brodley

ACM Transactions on Information and System Security (TISSEC) August 1999 Volume 2 Issue 3

The anomaly-detection problem can be formulated as one of learning to characterize the behaviors of an individual, system, or network in terms of temporal sequences of discrete data. We present an approach on the basis of instance-based learning (IBL) techniques. To cast the anomaly-detection task in an IBL framework, we employ an approach that transforms temporal sequences of discrete, unordered observations into a metric space via a similarity measure that encodes intra-attribute depende ...

30 User interactions with everyday applications as context for just-in-time information ৰী access Jay Budzik , Kristian J. Hammond Proceedings of the 5th international conference on Intelligent user interfaces January 2000 Our central claim is that user interactions with everyday productivity applications (e.g., word processors, Web browsers, etc.) provide rich contextual information that can be leveraged to support just-in-time access to task-relevant information. We discuss the requirements for such systems, and develop a general architecture for systems of this type. As evidence for our claim, we present Watson, a system which gathers contextual information in the form of the text of the document the user ... 31 The FINITE STRING Newsletter: Abstracts of current literature 77 Computational Linguistics Staff Computational Linguistics January 1987 Volume 13 Issue 1-2 77 32 The FINITE STRING newsletter: Abstracts of current literature Computational Linguistics Staff Computational Linguistics April 1986 Volume 12 Issue 2 33 Challenges in information retrieval and language modeling: report of a workshop 77 held at the center for intelligent information retrieval, University of Massachusetts Amherst, September 2002 James Allan , Jay Aslam , Nicholas Belkin , Chris Buckley , Jamie Callan , Bruce Croft , Sue Dumais , Norbert Fuhr , Donna Harman , David J. Harper , Djoerd Hiemstra , Thomas Hofmann , Eduard Hovy , Wessel Kraaij , John Lafferty , Victor Lavrenko , David Lewis , Liz Liddy , R. Manmatha , Andrew McCallum , Jay Ponte , John Prager , Dragomir Radev , Philip Resnik , Stephen Robertson , Roni Rosenfeld, Salim Roukos, Mark Sanderson, Rich Schwartz, Amit Singhal, Alan Smeaton, Howard Turtle, Ellen Voorhees, Ralph Weischedel, Jinxi Xu, ChengXiang Zhai **ACM SIGIR Forum** April 2003 Volume 37 Issue 1 34 Evolving data mining into solutions for insights: Scaling mining algorithms to large 77 databases Paul Bradley, Johannes Gehrke, Raghu Ramakrishnan, Ramakrishnan Srikant Communications of the ACM August 2002 Volume 45 Issue 8 Which insights about data structure make it possible to analyze the very large databases collected by Internet, business, scientific, and government applications? 35 Description and Analysis: ChangeDetector™: a site-level monitoring tool for the 77 A WWW Vijay Boyapati, Kristie Chevrier, Avi Finkel, Natalie Glance, Tom Pierce, Robert Stockton, Chip Whitmer Proceedings of the eleventh international conference on World Wide Web May 2002 This paper presents a new challenge for Web monitoring tools: to build a system that can monitor entire web sites effectively. Such a system could potentially be used to discover "silent news" hidden within corporate web sites. Examples of silent news include reorganizations in the executive team of a company or in the retirement of a product line. ChangeDetector, an implemented prototype,

addresses this challenge by incorporating a number of machine learning techniques. The principal

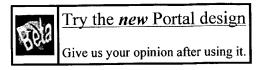
backend co ...

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40 4	Interactive two-handed gesture interface in 3D virtual environments Hiroaki Nishino , Kouichi Utsumiya , Daisuke Kuraoka , Kenji Yoshioka , Kazuyoshi Korida Proceedings of the ACM symposium on Virtual reality software and technology September 1997	77			
39 [4]	Programming by demonstration: an inductive learning formulation Tessa A. Lau, Daniel S. Weld Proceedings of the 4th international conference on Intelligent user interfaces December 1998	77			
38 4	A learning agent for wireless news access Daniel Billsus, Michael J. Pazzani, James Chen Proceedings of the 5th international conference on Intelligent user interfaces January 2000 We describe a user interface for wireless information devices, specifically designed to facilitate learning about users' individual interests in daily news stories. User feedback is collected unobtrusively to form the basis for a content-based machine learning algorithm. As a result, the described system can adapt to users' individual interests, reduce the amount of information that needs to be transmitted, and help users access relevant information with minimal effort.	77			
	The new (1982) Computing Reviews classification system—final version Jean E. Sammet , Anthony Ralston Communications of the ACM January 1982 Volume 25 Issue 1	77			
	The proposed new Computing Reviews classification scheme Anthony Ralston Communications of the ACM July 1981 Volume 24 Issue 7				



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41 4	Detection of shifts in user interests for personalized information filtering W. Lam , S. Mukhopadhyay , J. Mostafa , M. Palakal Proceedings of the 19th annual international ACM SIGIR conference on Research and development in information retrieval August 1996	77							
42 4	Mining scientific data Usama Fayyad , David Haussler , Paul Stolorz Communications of the ACM November 1996 Volume 39 Issue 11	77							
43 4	A multiparadigmatic environment for interacting with databases T. Catarci , M. F. Costabile , A. Massari , L. Saladini , G. Santucci ACM SIGCHI Bulletin July 1996 Volume 28 Issue 3 We present a prototype system to be used for visually accessing heterogeneous databases. The basic idea is to provide the user with several visual representations of data as well as multiple interaction mechanisms for both querying databases and visualizing the query results. Since some visual representations better fit certain user classes, the system adapts to the user's needs by switching to the most appropriate visual representation and interaction mechanism, according to a suitable user mod	77							

44 Pen computing: a technology overview and a vision

André Meyer

ACM SIGCHI Bulletin July 1995

Volume 27 Issue 3

This work gives an overview of a new technology that is attracting growing interest in public as well as in the computer industry itself. The visible difference from other technologies is in the use of a pen or pencil as the primary means of interaction between a user and a machine, picking up the familiar pen and paper interface metaphor. From this follows a set of consequences that will be analyzed and put into context with other emerging technologies and visions. Starting with a short historic ...

77

45 Automated cataloging and analysis of sky survey image databases: the SKICAT

system

Usama M. Fayyad , Nicholas Weir , S. Djorgovski

Proceedings of the second international conference on Information and knowledge management December 1993

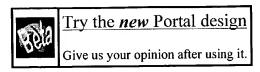
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Technique for automatically correcting words in text

87

Karen Kukich

ACM Computing Surveys (CSUR) December 1992

Volume 24 Issue 4

Research aimed at correcting words in text has focused on three progressively more difficult problems:(1) nonword error detection; (2) isolated-word error correction; and (3) context-dependent work correction. In response to the first problem, efficient pattern-matching and n-gram analysis techniques have been developed for detecting strings that do not appear in a given word list. In response to the second problem, a variety of general and application-specific spelling cor ...

Learning classifiers: Liveclassifier: creating hierarchical text classifiers through web 80 corpora

Chien-Chung Huang , Shui-Lung Chuang , Lee-Feng Chien

Proceedings of the 13th conference on World Wide Web May 2004

Many Web information services utilize techniques of information extraction(IE) to collect important facts from the Web. To create more advanced services, one possible method is to discover thematic information from the collected facts through text classification. However, most conventional text classification techniques rely on manual-labelled corpora and are thus ill-suited to cooperate with Web information services with open domains. In this work, we present a system named LiveClassifier that ...

Computational models: Biologically inspired rule-based multiset programming paradigm for soft-computing

80

E. V. Krishnamurthy, V. K. Murthy, Vikram Krishnamurthy

Proceedings of the first conference on computing frontiers on Computing frontiers April 2004 This paper describes a rule-based multiset programming paradigm, as a unifying theme for biological. chemical, DNA, physical and molecular computations. The computations are interpreted as the outcome arising out of deterministic, nondeterministic or stochastic interaction among elements in a multiset object space which includes the environment. These interactions are like chemical reactions and the evolution of the multiset can mimic the biological evolution. Since the reaction rules are inhere ...

4 Ontological user profiling in recommender systems

Stuart E. Middleton , Nigel R. Shadbolt , David C. De Roure

ACM Transactions on Information Systems (TOIS) January 2004

Volume 22 Issue 1

We explore a novel ontological approach to user profiling within recommender systems, working on the problem of recommending on-line academic research papers. Our two experimental systems, Quickstep and Foxtrot, create user profiles from unobtrusively monitored behaviour and relevance feedback, representing the profiles in terms of a research paper topic ontology. A novel profile visualization approach is taken to acquire profile feedback. Research papers are classified using ontological classes ...

5 A model of multimedia information retrieval

80

Carlo Meghini , Fabrizio Sebástiani , Umberto Straccia **Journal of the ACM (JACM)** September 2001

Volume 48 Issue 5

Research on multimedia information retrieval (MIR) has recently witnessed a booming interest. A prominent feature of this research trend is its simultaneous but independent materialization within several fields of computer science. The resulting richness of paradigms, methods and systems may, on the long run, result in a fragmentation of efforts and slow down progress. The primary goal of this study is to promote an integration of methods and techniques for MIR by contributing a conceptual model ...

6 Data clustering: a review

80

A. K. Jain , M. N. Murty , P. J. Flynn

ACM Computing Surveys (CSUR) September 1999

Volume 31 Issue 3

Clustering is the unsupervised classification of patterns (observations, data items, or feature vectors) into groups (clusters). The clustering problem has been addressed in many contexts and by researchers in many disciplines; this reflects its broad appeal and usefulness as one of the steps in exploratory data analysis. However, clustering is a difficult problem combinatorially, and differences in assumptions and contexts in different communities has made the transfer of useful generic co ...

7 Knowledge and representation: Leveraging a common representation for personalized search and summarization in a medical digital library Kathleen R. McKeown, Noemie Elhadad, Vasileios Hatzivassiloglou

80

Proceedings of the third ACM/IEEE-CS joint conference on Digital libraries May 2003

Despite the large amount of online medical literature, it can be difficult for clinicians to find relevant information at the point of patient care. In this paper, we present techniques to personalize the results of search, making use of the online patient record as a sophisticated, pre-existing user model. Our work in *PERSIVAL*, a medical digital library, includes methods for re-ranking the results of search to prioritize those that better match the patient record. It also generates summa ...

8 Efficient algorithms for geometric optimization

80

Pankaj K. Agarwal , Micha Sharir

ACM Computing Surveys (CSUR) December 1998

Volume 30 Issue 4

We review the recent progress in the design of efficient algorithms for various problems in geometric optimization. We present several techniques used to attack these problems, such as parametric searching, geometric alternatives to parametric searching, prune-and-search techniques for linear programming and related problems, and LP-type problems and their efficient solution. We then describe a wide range of applications of these and other techniques to numerous problems in geometric optim ...

9 Supporting cooperative and personal surfing with a desktop assistant

Hannes Marais , Krishna Bharat

Proceedings of the 10th annual ACM symposium on User interface software and technology October 1997

10 A multilevel approach to intelligent information filtering: model, system, and

80

4 evaluation

J. Mostafa , S. Mukhopadhyay , M. Palakal , W. Lam

ACM Transactions on Information Systems (TOIS) October 1997

Volume 15 Issue 4

In information-filtering environments, uncertainties associated with changing interests of the user and the dynamic document stream must be handled efficiently. In this article, a filtering model is proposed that decomposes the overall task into subsystem functionalities and highlights the need for multiple adaptation techniques to cope with uncertainties. A filtering system, SIFTER, has been implemented based on the model, using established techniques in information retrieval and artificia ...

11 Semantic annotation and integration: Towards the self-annotating web

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16 Data streams (DS): Discovering decision rules from numerical data streams

77

Francisco Ferrer-Troyano , Jesús S. Aguilar-Ruiz , José C. Riquelme

Proceedings of the 2004 ACM symposium on Applied computing March 2004

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18 Phase tracking and prediction

77

Timothy Sherwood , Suleyman Sair , Brad Calder

ACM SIGARCH Computer Architecture News, Proceedings of the 30th annual international symposium on Computer architecture May 2003

Volume 31 Issue 2

In a single second a modern processor can execute billions of instructions. Obtaining a bird's eve view of the behavior of a program at these speeds can be a difficult task when all that is available is cycle by cycle examination. In many programs, behavior is anything but steady state, and understanding the patterns of behavior, at run-time, can unlock a multitude of optimization opportunities. In this paper, we present a unified profiling architecture that can efficiently capture, classify, and ...

19 Survey articles: Web usage mining: discovery and applications of usage patterns

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Jaideep Srivastava , Robert Cooley , Mukund Deshpande , Pang-Ning Tan

ACM SIGKDD Explorations Newsletter January 2000

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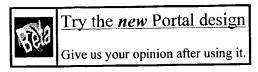
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23 Machine learning in automated text categorization

Fabrizio Sebastiani
ACM Computing Surveys (CSUR) March 2002

Volume 34 Issue 1

The automated categorization (or classification) of texts into predefined categories has witnessed a booming interest in the last 10 years, due to the increased availability of documents in digital form and the ensuing need to organize them. In the research community the dominant approach to this problem is based on machine learning techniques: a general inductive process automatically builds a classifier by learning, from a set of preclassified documents, the characteristics of the categories. ...

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77

77

Cody Kwok , Oren Etzioni , Daniel S. Weld ACM Transactions on Information Systems (TOIS) July 2001 Volume 19 Issue 3 The wealth of information on the web makes it an attractive resource for seeking quick answers to simple, factual questions such as "e; who was the first American in space?"e; or "e; what is the second tallest mountain in the world?"e; Yet today's most advanced web search services (e.g., Google and AskJeeves) make it surprisingly tedious to locate answers to such questions. In this paper, we extend question-answering techniques, first studied in the information retrieval literature ... 25 Video Retrieval and Browsing: Comparing discriminating transformations and SVM for learning during multimedia retrieval Xiang Sean Zhou, Thomas S. Huang On-line learning or "relevance feedback" techniques for multimedia information retrieval have been

Proceedings of the ninth ACM international conference on Multimedia October 2001

explored from many different points of view: from early heuristic-based feature weighting schemes to recently proposed optimal learning algorithms, probabilistic/Bayesian learning algorithms, boosting techniques, discriminant-EM algorithm, support vector machine, and other kernel-based learning machines. Based on a careful examination of the problem and a detailed analysis of the existing solutions ...

26 Scaling question answering to the Web Cody C. T. Kwok , Oren Etzioni , Daniel S. Weld

Proceedings of the tenth international conference on World Wide Web April 2001

27 Temporal sequence learning and data reduction for anomaly detection

Terran Lane , Carla E. Brodlev ACM Transactions on Information and System Security (TISSEC) August 1999 Volume 2 Issue 3

The anomaly-detection problem can be formulated as one of learning to characterize the behaviors of an individual, system, or network in terms of temporal sequences of discrete data. We present an approach on the basis of instance-based learning (IBL) techniques. To cast the anomaly-detection task in an IBL framework, we employ an approach that transforms temporal sequences of discrete, unordered observations into a metric space via a similarity measure that encodes intra-attribute depende ...

28 User interactions with everyday applications as context for just-in-time information

access Jav Budzik , Kristian J. Hammond

Proceedings of the 5th international conference on Intelligent user interfaces January 2000 Our central claim is that user interactions with everyday productivity applications (e.g., word processors, Web browsers, etc.) provide rich contextual information that can be leveraged to support just-in-time access to task-relevant information. We discuss the requirements for such systems, and develop a general architecture for systems of this type. As evidence for our claim, we present Watson, a system which gathers contextual information in the form of the text of the document the user ...

29 The FINITE STRING Newsletter: Abstracts of current literature

Computational Linguistics Staff Computational Linguistics January 1987 Volume 13 Issue 1-2

30 The FINITE STRING newsletter: Abstracts of current literature

Computational Linguistics Staff Computational Linguistics April 1986

Volume 12 Issue 2

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31 4	Challenges in information retrieval and language modeling: report of a workshop held at the center for intelligent information retrieval, University of Massachusetts Amherst, September 2002 James Allan , Jay Aslam , Nicholas Belkin , Chris Buckley , Jamie Callan , Bruce Croft , Sue Dumais , Norbert Fuhr , Donna Harman , David J. Harper , Djoerd Hiemstra , Thomas Hofmann , Eduard Hovy , Wessel Kraaij , John Lafferty , Victor Lavrenko , David Lewis , Liz Liddy , R. Manmatha , Andrew McCallum , Jay Ponte , John Prager , Dragomir Radev , Philip Resnik , Stephen Robertson , Roni Rosenfeld , Salim Roukos , Mark Sanderson , Rich Schwartz , Amit Singhal , Alan Smeaton , Howard Turtle , Ellen Voorhees , Ralph Weischedel , Jinxi Xu , ChengXiang Zhai ACM SIGIR Forum April 2003 Volume 37 Issue 1	77
32 4	Description and Analysis: ChangeDetector™: a site-level monitoring tool for the WWW Vijay Boyapati , Kristie Chevrier , Avi Finkel , Natalie Glance , Tom Pierce , Robert Stockton , Chip Whitmer Proceedings of the eleventh international conference on World Wide Web May 2002 This paper presents a new challenge for Web monitoring tools: to build a system that can monitor entire web sites effectively. Such a system could potentially be used to discover "silent news" hidden within corporate web sites. Examples of silent news include reorganizations in the executive team of a company or in the retirement of a product line. ChangeDetector, an implemented prototype, addresses this challenge by incorporating a number of machine learning techniques. The principal backend co	77
33 4	The proposed new Computing Reviews classification scheme Anthony Ralston Communications of the ACM July 1981 Volume 24 Issue 7	77
34 4	The new (1982) Computing Reviews classification system—final version Jean E. Sammet, Anthony Ralston Communications of the ACM January 1982 Volume 25 Issue 1	77
35 4	A learning agent for wireless news access Daniel Billsus, Michael J. Pazzani, James Chen Proceedings of the 5th international conference on Intelligent user interfaces January 2000 We describe a user interface for wireless information devices, specifically designed to facilitate learning about users' individual interests in daily news stories. User feedback is collected unobtrusively to form the basis for a content-based machine learning algorithm. As a result, the described system can adapt to users' individual interests, reduce the amount of information that needs to be transmitted, and help users access relevant information with minimal effort.	77
30 _4	Detection of shifts in user interests for personalized information filtering W. Lam , S. Mukhopadhyay , J. Mostafa , M. Palakal Proceedings of the 19th annual international ACM SIGIR conference on Research and development in information retrieval August 1996	77
37 [∢	7 Mining scientific data Usama Fayyad , David Haussler , Paul Stolorz Communications of the ACM November 1996 Volume 39 Issue 11	77

38 A multiparadigmatic environment for interacting with databases

T. Catarci , M. F. Costabile , A. Massari , L. Saladini , G. Santucci

ACM SIGCHI Bulletin July 1996

Volume 28 Issue 3

We present a prototype system to be used for visually accessing heterogeneous databases. The basic idea is to provide the user with several visual representations of data as well as multiple interaction mechanisms for both querying databases and visualizing the query results. Since some visual representations better fit certain user classes, the system adapts to the user's needs by switching to the most appropriate visual representation and interaction mechanism, according to a suitable user mod ...

39 Pen computing: a technology overview and a vision

77

André Meyer

ACM SIGCHI Bulletin July 1995

Volume 27 Issue 3

This work gives an overview of a new technology that is attracting growing interest in public as well as in the computer industry itself. The visible difference from other technologies is in the use of a pen or pencil as the primary means of interaction between a user and a machine, picking up the familiar pen and paper interface metaphor. From this follows a set of consequences that will be analyzed and put into context with other emerging technologies and visions. Starting with a short historic ...

40 Automated cataloging and analysis of sky survey image databases: the SKICAT

77

🐴 system

Usama M. Fayyad , Nicholas Weir , S. Djorgovski

Proceedings of the second international conference on Information and knowledge management December 1993

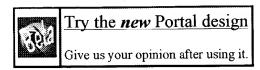
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1 Survey articles: Data mining for hypertext: a tutorial survey

87%

Soumen Chakrabarti

ACM SIGKDD Explorations Newsletter January 2000

Volume 1 Issue 2

With over 800 million pages covering most areas of human endeavor, the World-wide Web is a fertile ground for data mining research to make a difference to the effectiveness of information search. Today, Web surfers access the Web through two dominant interfaces: clicking on hyperlinks and searching via keyword queries. This process is often tentative and unsatisfactory. Better support is needed for expressing one's information need and dealing with a search result in more structured ways than av ...

2 Selective sampling for example-based word sense disambiguation

85%

Atsushi Fujii , Takenobu Tokunaga , Kentaro Inui , Hozumi Tanaka Computational Linguistics December 1998

Volume 24 Issue 4

This paper proposes an efficient example sampling method for example-based word sense disambiguation systems. To construct a database of practical size, a considerable overhead for manual sense disambiguation (overhead for supervision) is required. In addition, the time complexity of searching a large-sized database poses a considerable problem (overhead for search). To counter these problems, our method selectively samples a smaller-sized effective subset from a given example set for use in wor ...

3 Machine learning in automated text categorization

82%

Fabrizio Sebastiani

ACM Computing Surveys (CSUR) March 2002

Volume 34 Issue 1

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classifier by learning, from a set of preclassified documents, the characteristics of the categories. ...

Learning classifiers: Liveclassifier: creating hierarchical text classifiers through web 82%

Chien-Chung Huang, Shui-Lung Chuang, Lee-Feng Chien

Proceedings of the 13th conference on World Wide Web May 2004

Many Web information services utilize techniques of information extraction(IE) to collect important facts from the Web. To create more advanced services, one possible method is to discover thematic information from the collected facts through text classification. However, most conventional text classification techniques rely on manual-labelled corpora and are thus ill-suited to cooperate with Web information services with open domains. In this work, we present a system named LiveClassifier that ...

Data clustering: a review

82%

A. K. Jain, M. N. Murty, P. J. Flynn

ACM Computing Surveys (CSUR) September 1999

Volume 31 Issue 3

Clustering is the unsupervised classification of patterns (observations, data items, or feature vectors) into groups (clusters). The clustering problem has been addressed in many contexts and by researchers in many disciplines; this reflects its broad appeal and usefulness as one of the steps in exploratory data analysis. However, clustering is a difficult problem combinatorially, and differences in assumptions and contexts in different communities has made the transfer of useful generic co ...

Exploration of text collections with hierarchical feature maps

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Dieter Merkl

ACM SIGIR Forum, Proceedings of the 20th annual international ACM SIGIR conference on Research and development in information retrieval July 1997

Volume 31 Issue SI

Special issue on word sense disambiguation: Disambiguating highly ambiguous

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Geoffrey Towell, Ellen M. Voorhees

Computational Linguistics March 1998

Volume 24 Issue 1

A word sense disambiguator that is able to distinguish among the many senses of common words that are found in general-purpose, broad-coverage lexicons would be useful. For example, experiments have shown that, given accurate sense disambiguation, the lexical relations encoded in lexicons such as WordNet can be exploited to improve the effectiveness of information retrieval systems. This paper describes a classifier whose accuracy may be sufficient for such a purpose. The classifier combines the ...

Special issue on special feature: Sufficient dimensionality reduction

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Amir Globerson , Naftali Tishby

The Journal of Machine Learning Research March 2003

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Enhanced hypertext categorization using hyperlinks

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Soumen Chakrabarti , Byron Dom , Piotr Indyk

ACM SIGMOD Record, Proceedings of the 1998 ACM SIGMOD international conference on Management of data June 1998

Volume 27 Issue 2

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10 Automated techniques for managing collections: Machine learning for information

77%

architecture in a large governmental website

Miles Efron , Jonathan Elsas , Gary Marchionini , Junliang Zhang

Proceedings of the 2004 joint ACM/IEEE conference on Digital libraries June 2004

This paper describes ongoing research into the application of machine learning techniques for improving access to governmental information in complex digital libraries. Under the auspices of the GovStat Project, our goal is to identify a small number of semantically valid concepts that adequately spans the intellectual domain of a collection. The goal of this discovery is twofold. First we desire a practical aid for information architects. Second, automatically derived document-concept relations ...

11 Word sense disambiguation of adjectives using probabilistic networks

77%

Gerald Chao , Michael G. Dyer

Proceedings of the 17th conference on Computational linguistics - Volume 1 July 2000 In this paper, word sense disambiguation (WSD) accuracy achievable by a probabilistic classifier, using very minimal training sets, is investigated. We made the assumption that there are no tagged corpora available and identified what information, needed by an accurate WSD system, can and cannot be automatically obtained. The lesson learned can then be used to focus on what knowledge needs manual annotation. Our system, named Bayesian Hierarchical Disambiguator (BHD), uses the Internet, a ...

12 Image retrieval: A bootstrapping approach to annotating large image collection

77%

HuaMin Feng , Tat-Seng Chua

Proceedings of the 5th ACM SIGMM international workshop on Multimedia information retrieval November 2003

Huge amount of manual efforts are required to annotate large image/video archives with text annotations. Several recent works attempted to automate this task by employing supervised learning approaches to associate visual information extracted in segmented images with semantic concepts provided by associated text. The main limitation of such approaches, however, is that large labeled training corpus is still needed for effective learning, and semantically meaningful segmentation for images is in ...

13 Special issue on word sense disambiguation: Topical clustering of MRD senses based 77%

4 on information retrieval techniques

Jen Nan Chen, Jason S. Chang

Computational Linguistics March 1998

Volume 24 Issue 1

This paper describes a heuristic approach capable of automatically clustering senses in a machinereadable dictionary (MRD). Including these clusters in the MRD-based lexical database offers several positive benefits for word sense disambiguation (WSD). First, the clusters can be used as a coarser sense division, so unnecessarily fine sense distinction can be avoided. The clustered entries in the MRD can also be used as materials for supervised training to develop a WSD system. Furthermore, if t ...

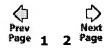
77% 14 Special issue on word sense disambiguation: Introduction to the special issue on word sense disambiguation: the state of the art Nancy Ide , Jean Véronis Computational Linguistics March 1998 Volume 24 Issue 1 77% 15 Video Retrieval and Browsing: Comparing discriminating transformations and SVM A) for learning during multimedia retrieval Xiang Sean Zhou, Thomas S. Huang Proceedings of the ninth ACM international conference on Multimedia October 2001 On-line learning or "relevance feedback" techniques for multimedia information retrieval have been explored from many different points of view: from early heuristic-based feature weighting schemes to recently proposed optimal learning algorithms, probabilistic/Bayesian learning algorithms, boosting techniques, discriminant-EM algorithm, support vector machine, and other kernel-based learning machines. Based on a careful examination of the problem and a detailed analysis of the existing solutions ... 77% 16 Adaptive information filtering: detecting changes in text streams Carsten Languillon , Ingrid Renz Proceedings of the eighth international conference on Information and knowledge management November 1999 The task of information filtering is to classify documents from a stream as either relevant or nonrelevant according to a particular user interest with the objective to reduce information load. When using an information filter in an environment that is changing with time, methods for adapting the filter should be considered in order to retain classification accuracy. We favor a methodology that attempts to detect changes and adapts the information filter only if inevitable in order to mini ... 77% 17 Content-based book recommending using learning for text categorization Raymond J. Mooney, Loriene Roy Proceedings of the fifth ACM conference on Digital libraries June 2000 Recommender systems improve access to relevant products and information by making personalized suggestions based on previous examples of a user's likes and dislikes. Most existing recommender systems use collaborative filtering methods that base recommendations on other users' preferences. By contrast, content-based methods use information about an item itself to make suggestions. This approach has the advantage of being able to recommend previously unrated items to users with unique interes ... 77% 18 Learnable visual keywords for image classification Proceedings of the fourth ACM conference on Digital libraries August 1999 77% 19 Web mining research: a survey Raymond Kosala , Hendrik Blockeel **ACM SIGKDD Explorations Newsletter** June 2000 Volume 2 Issue 1 20 Knowledge management session 4: indexing: Bootstrapping for hierarchical 77% বী document classification Giordano Adami , Paolo Avesani , Diego Sona Proceedings of the twelfth international conference on Information and knowledge management November 2003

Managing the hierarchical organization of data is starting to play a key role in the knowledge

management community due to the great amount of human resources needed to create and maintain these organized repositories of information. Machine learning community has in part addressed this problem by developing hierarchical supervised classifiers that help maintainers to categorize new resources within given hierarchies. Although such learning models succeed in exploiting relational knowledge, they ...

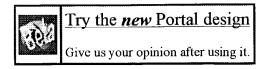
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21 Challenges in information retrieval and language modeling: report of a workshop held at the center for intelligent information retrieval, University of Massachusetts Amherst, September 2002

James Allan , Jay Aslam , Nicholas Belkin , Chris Buckley , Jamie Callan , Bruce Croft , Sue Dumais , Norbert Fuhr, Donna Harman, David J. Harper, Djoerd Hiemstra, Thomas Hofmann, Eduard Hovy, Wessel Kraaij , John Lafferty , Victor Lavrenko , David Lewis , Liz Liddy , R. Manmatha , Andrew McCallum , Jay Ponte , John Prager , Dragomir Radev , Philip Resnik , Stephen Robertson , Roni Rosenfeld, Salim Roukos, Mark Sanderson, Rich Schwartz, Amit Singhal, Alan Smeaton, Howard Turtle, Ellen Voorhees, Ralph Weischedel, Jinxi Xu, ChengXiang Zhai

ACM SIGIR Forum April 2003

Volume 37 Issue 1

22 Special issue on Machine learning methods for text and images: Matching words and 77% বা pictures

Kobus Barnard, Pinar Duygulu, David Forsyth, Nando de Freitas, David M. Blei, Michael I. Jordan The Journal of Machine Learning Research March 2003 Volume 3

We present a new approach for modeling multi-modal data sets, focusing on the specific case of segmented images with associated text. Learning the joint distribution of image regions and words has many applications. We consider in detail predicting words associated with whole images (autoannotation) and corresponding to particular image regions (region naming). Auto-annotation might help organize and access large collections of images. Region naming is a model of object recognition as a process ...

23 Learning with mixtures of trees

Marina Meila, Michael I. Jordan

The Journal of Machine Learning Research September 2001

Volume 1

This paper describes the mixtures-of-trees model, a probabilistic model for discrete multidimensional

77%

domains. Mixtures-of-trees generalize the probabilistic trees of Chow and Liu (1968) in a different and complementary direction to that of Bayesian networks. We present efficient algorithms for learning mixtures-of-trees models in maximum likelihood and Bayesian frameworks. We also discuss additional efficiencies that can be obtained when data are "sparse," and we present data structures and alg ...

24 A comparative study for domain ontology guided feature extraction

77%

Bill B. Wang, R. I. Bob Mckay, Hussein A. Abbass, Michael Barlow

Proceedings of the twenty-sixth Australasian computer science conference on Conference in research and practice in information technology - Volume 16 February 2003

We introduced a novel method employing a hierarchical domain ontology structure to extract features representing documents in our previous publication (Wang 2002). All raw words in the training documents are mapped to concepts in a concept hierarchy derived from the domain ontology. Based on these concepts, a concept hierarchy is established for the training document space, using is-a relationships defined in the domain ontology. An optimum concept set may be obtained by searching the concept hi ...

25 Scalable feature selection, classification and signature generation for organizing

77%

A large text databases into hierarchical topic taxonomies

 $Soumen\ Chakrabarti\ ,\ Byron\ Dom\ ,\ Rakesh\ Agrawal\ ,\ Prabhakar\ Raghavan$

The VLDB Journal — The International Journal on Very Large Data Bases August 1998 Volume 7 Issue 3

We explore how to organize large text databases hierarchically by topic to aid better searching, browsing and filtering. Many corpora, such as internet directories, digital libraries, and patent databases are manually organized into topic hierarchies, also called *taxonomies*. Similar to indices for relational data, taxonomies make search and access more efficient. However, the exponential growth in the volume of on-line textual information makes it nearly impossible to maintain such taxono ...

26 Summarization: The use of unlabeled data to improve supervised learning for text

77%

4 summarization

Massih-Reza Amini , Patrick Gallinari

Proceedings of the 25th annual international ACM SIGIR conference on Research and development in information retrieval August 2002

With the huge amount of information available electronically, there is an increasing demand for automatic text summarization systems. The use of machine learning techniques for this task allows one to adapt summaries to the user needs and to the corpus characteristics. These desirable properties have motivated an increasing amount of work in this field over the last few years. Most approaches attempt to generate summaries by extracting sentence segments and adopt the supervised learning paradigm ...

27 Fast supervised dimensionality reduction algorithm with applications to document

77%

d categorization & retrieval

George Karypis, Eui-Hong (Sam) Han

Proceedings of the ninth international conference on Information and knowledge management
November 2000

28 Hypertext data mining (tutorial AM-1)

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Soumen Chakrabarti

Tutorial notes of the sixth ACM SIGKDD international conference on Knowledge discovery and data mining August 2000

29 Classification and regression: money *can* grow on trees

4

Johannes Gehrke, Wie-Yin Loh, Raghu Ramakrishnan

Tutorial notes of the fifth ACM SIGKDD international conference on Knowledge discovery and data mining August 1999

With over 800 million pages covering most areas of human endeavor, the World-wide Web is a fertile ground for data mining research to make a difference to the effectiveness of information search. Today, Web surfers access the Web through two dominant interfaces clicking on hyperlinks and searching via keyword queries This process is often tentative and unsatisfactory Better support is needed for expressing one's information need and dealing with a search result in more structured ways than ...

30 An evaluation of phrasal and clustered representations on a text categorization task 77%

David D. Lewis

Proceedings of the 15th annual international ACM SIGIR conference on Research and development in information retrieval June 1992

Syntactic phrase indexing and term clustering have been widely explored as text representation techniques for text retrieval. In this paper we study the properties of phrasal and clustered indexing languages on a text categorization task, enabling us to study their properties in isolation from query interpretation issues. We show that optimal effectiveness occurs when using only a small proportion of the indexing terms available, and that effectiveness peaks at a higher feature set size and ...

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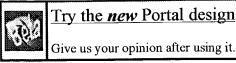
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Soumen Chakrabarti

ACM SIGKDD Explorations Newsletter January 2000

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Computational Linguistics March 1998

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Amir Globerson, Naftali Tishby

The Journal of Machine Learning Research March 2003

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Dieter Merkl

ACM SIGIR Forum , Proceedings of the 20th annual international ACM SIGIR conference on Research and development in information retrieval July 1997

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11 Special issue on word sense disambiguation: Introduction to the special issue on

word sense disambiguation: the state of the art

Nancy Ide, Jean Véronis

Computational Linguistics March 1998

Volume 24 Issue 1

12 Video Retrieval and Browsing: Comparing discriminating transformations and SVM

for learning during multimedia retrieval

Xiang Sean Zhou, Thomas S. Huang

Proceedings of the ninth ACM international conference on Multimedia October 2001

On-line learning or "relevance feedback" techniques for multimedia information retrieval have been explored from many different points of view: from early heuristic-based feature weighting schemes to recently proposed optimal learning algorithms, probabilistic/Bayesian learning algorithms, boosting techniques, discriminant-EM algorithm, support vector machine, and other kernel-based learning machines. Based on a careful examination of the problem and a detailed analysis of the existing solutions ...

13 Adaptive information filtering: detecting changes in text streams

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Carsten Lanquillon , Ingrid Renz

Proceedings of the eighth international conference on Information and knowledge

management November 1999

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14 Content-based book recommending using learning for text categorization

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Raymond J. Mooney , Loriene Roy

Proceedings of the fifth ACM conference on Digital libraries June 2000

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15 Challenges in information retrieval and language modeling: report of a workshop held at the center for intelligent information retrieval, University of Massachusetts Amherst, September 2002

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ACM SIGIR Forum April 2003

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16 Special issue on Machine learning methods for text and images: Matching words and 77% pictures

Kobus Barnard , Pinar Duygulu , David Forsyth , Nando de Freitas , David M. Blei , Michael I. Jordan **The Journal of Machine Learning Research** March 2003

Volume 3

We present a new approach for modeling multi-modal data sets, focusing on the specific case of segmented images with associated text. Learning the joint distribution of image regions and words has many applications. We consider in detail predicting words associated with whole images (auto-annotation) and corresponding to particular image regions (region naming). Auto-annotation might help organize and access large collections of images. Region naming is a model of object recognition as a process ...

17 Learning with mixtures of trees

77%

Marina Meila , Michael I. Jordan

The Journal of Machine Learning Research September 2001

Volume 1

This paper describes the mixtures-of-trees model, a probabilistic model for discrete multidimensional domains. Mixtures-of-trees generalize the probabilistic trees of Chow and Liu (1968) in a different and complementary direction to that of Bayesian networks. We present efficient algorithms for learning mixtures-of-trees models in maximum likelihood and Bayesian frameworks. We also discuss additional efficiencies that can be obtained when data are "sparse," and we present data structures and alg ...

18 Scalable feature selection, classification and signature generation for organizing

77%

large text databases into hierarchical topic taxonomies

Soumen Chakrabarti , Byron Dom , Rakesh Agrawal , Prabhakar Raghavan

The VLDB Journal — The International Journal on Very Large Data Bases August 1998 Volume 7 Issue 3

We explore how to organize large text databases hierarchically by topic to aid better searching, browsing and filtering. Many corpora, such as internet directories, digital libraries, and patent databases are manually organized into topic hierarchies, also called *taxonomies*. Similar to indices for relational data, taxonomies make search and access more efficient. However, the exponential growth in the volume of on-line textual information makes it nearly impossible to maintain such taxono ...

19 Summarization: The use of unlabeled data to improve supervised learning for text

Massih-Reza Amini , Patrick Gallinari

Proceedings of the 25th annual international ACM SIGIR conference on Research and development in information retrieval August 2002

With the huge amount of information available electronically, there is an increasing demand for automatic text summarization systems. The use of machine learning techniques for this task allows one to adapt summaries to the user needs and to the corpus characteristics. These desirable properties have motivated an increasing amount of work in this field over the last few years. Most approaches attempt to generate summaries by extracting sentence segments and adopt the supervised learning paradigm ...

20 Fast supervised dimensionality reduction algorithm with applications to document

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d categorization & retrieval

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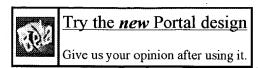
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